

Spatial patterns of climate-growth relationships across species distribution as a forest management tool in Moncayo Natural Park (Spain)

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Abstract

Forests exhibit strategies to cope with climate change; however, the rate of the changes on forests can be slower than the actual changes in environmental conditions. Forest management policies, such as assisted migration, may help forests to adapt their species distribution to changing climate conditions. Nonetheless, it certainly requires a better knowledge of climate influences on trees to ensure the success of specific management actions. In this study, we apply dendroclimatological methods to investigate the growth response of the main forest species present in Moncayo Natural Park to climate to assess their current relationship and to model these responses over the potential distribution of each species across the study area. Our results revealed large differences in the response of beech, pine and Pyrenean oak to prevailing climate factors and indicated species-specific patterns of climate sensitivity. The general importance of

summer conditions for tree growth was confirmed. In addition, we found directional trends in correlation with specific climate factors along spatial gradients; these results are consistent with the autoecology of the studied species. Based on these findings, we present a new model approach that can serve as a key tool for forest managers to design forest communities that are more stable during climatic change.

Keywords

Forest Management; Climate; Tree Growth; GAMs; Dendrochronology

1. INTRODUCTION

Climate is one of the main abiotic factors that defines and limits the natural distribution of tree species (Mott 2010). Therefore, it is expected that predicted climate change (IPCC 2013) will have a significant impact on the distribution of species (Pearson and Dawson 2003).

Tree species often exhibit physiological adaptations along climatic gradients, adapting to the specific climatic and site conditions (Sáenz-Romero et al. 2016). Alternatively, species can migrate in response to climate changes. However, both functional adaptation and migration are slow processes for long-lived organisms, such as trees, and likely do not keep pace with the actual change in environmental conditions (Sáenz-Romero et al. 2016).

Management of natural protected areas and commercial forests requires long-term strategies and practical guidelines to facilitate management decisions in response to climate change (Nabuurs et al. 2017; Walentowski et al. 2017; Loran et al. 2018). Assisted migration is an important tool to adjust species distributions to changing climate conditions and mitigate its effects. However, implementing assisted migration

46 requires knowledge of climate-growth relationships (CGRs) within a species' range to
47 ensure the success of the reforestation plan.

48 Tree-ring research is a powerful tool to assess the dynamics of climate-growth
49 relationships, including reactions to extreme climate events (Fritts 1972) throughout
50 long time periods. Using a tree-centered approach opens the possibility to understand
51 tree function under specific climate and site conditions (Sass-Klaassen et al. 2016). This
52 comes with the limitation that the gained information represents individual tree
53 populations growing under specific site conditions.

54 There are numerous studies demonstrating that CGRs can strongly vary along altitudinal
55 (Ponocná et al. 2016; Kharal et al. 2017) or environmental gradients (Čufar et al. 2014;
56 Kraus et al. 2016; Martínez del Castillo et al. 2018). This suggests a different ability of
57 species to cope with climate variability and change across their distribution area. De
58 Luis et al., (2013) presented an extensive tree-ring network for Aleppo pine (*Pinus*
59 *halepensis* Mill.) that was used to assess CGRs and trends in climate sensitivity across
60 the distribution area. Climate-growth relationships can be systematically assessed for
61 species along gradients through the distribution area of the species to systematically
62 extend this site-specific information.

63 It is difficult to obtain strong directional trends in climate response across the
64 distribution range because of the diversity in environments encountered across the
65 distribution area of a given species. As a consequence, there is a lack of information
66 about how tree species respond to climate conditions outside of their distribution range
67 and within the range (i.e., in areas of recent expansion, mixed stands or juvenile trees).

68 Interestingly, De Luis et al., (2013) demonstrated that such variation in CGRs are
69 related to prevailing climate conditions which allow for extrapolation of estimated
70 trends across the species distribution and also into areas where the species is currently

71 not present. Such information is useful for foresters that seek to perform assisted-
72 migration actions, specifically into sensitive areas, i.e., at the edge of the current
73 distribution area.

74 Moncayo Natural Park (MNP) is a mountain ecotone in a transitional zone between the
75 Eurosiberian and Mediterranean biogeographical region (Longares Aladrén 2004) and
76 forms a geographical boundary for several tree species. The ecotone areas are
77 transitional areas between two neighboring biomes that contain different vegetation
78 types and generally are more sensitive to environmental disturbances (Risser 1995).

79 MNP provides a relevant sample area with a number of different sites that span different
80 climatic types. Therefore, this area is a suitable location to apply new analysis
81 techniques aimed to spatially test the influence of climate on forest species.

82 European beech (*Fagus sylvatica* L.), Scots pine (*Pinus sylvestris* L.), and Mountain
83 pine (*Pinus uncinata* Ram.) are widespread European tree species with high economic
84 and ecological value. The forests stands in Moncayo Natural Park represent one of the
85 southern limits for all these species. Pyrenean oak (*Quercus pyrenaica* Willd.)
86 distribution in Europe is limited to the Atlantic climate areas of western France and the
87 Iberian Peninsula (Nieto Quintano et al. 2016); MNP is the northern extent of its
88 Mediterranean range. The current distribution of these species in this Natural Park are
89 influenced by historical land use, such as sheep and goat grazing and pine introduction
90 via reforestation (Arrechea 2002). Graphical records (e.g., old photographs) and aerial
91 photography indicate that the currently forested area was a meadow with isolated
92 patches of beech and Pyrenean oak trees about one hundred years ago. The closed-
93 canopy forest is relatively young, and the current situation does not reflect the potential
94 spatial distribution of the species across this site. Recently, several changes in forest
95 size and structure have been observed due to the abandonment of traditional uses

96 triggered by protection of the region as a natural park (Martínez del Castillo et al. 2015).
97 Specifically, beech and Pyrenean oak trees are protected by the Habitats
98 Directive (Council Directive 92/43/EEC).
99 In this study, we investigated the growth response of beech, Scots pine, Mountain pine
100 and Pyrenean oak to climate across an altitudinal-climatic gradient, and we modeled
101 these responses over the potential distribution of each species inside the Natural Park.
102 The following were our hypotheses: (i) the species exhibit different climate-growth
103 relationships, which vary along the climatic gradient; (ii) variation in climate-growth
104 relationships for each species across the altitudinal gradient is associated with the
105 specific climatic variability; and (iii) climate-growth relationships models can be
106 applied beyond the current distribution of each species to allocate potential migration
107 areas (either natural or artificial via reforestation). Our final aim was to predict CGRs
108 across the potential species distribution in MNP to investigate the use of this novel
109 approach for the design of forest management activities geared to adapt species
110 distributions to changing climate conditions. The obtained empirical model is based on
111 the specific local conditions at MNP and hence may not be widely applicable. However,
112 this new modelling approach illustrates the value of building models for specific areas
113 that can be applied by foresters to manage the distribution of species. Assuming that
114 tree secondary growth can be used as an appropriate indicator of tree performance
115 related with the environmental conditions, the interpretation of this results can provide
116 insights to desing management actions geared to create stable forest communities that
117 are able to cope with stress in changing (and more severe) climatic conditions.

2. METHODS

2.1. Study area

Moncayo Natural Park is a mountain area located in the Iberian range, in the northeast of the Iberian Peninsula (Fig. 1). This Natural Park covers 11,226 ha², and the elevation ranges from 850 m to 2314 m above sea level. It is located in a transitional climate area between the Mediterranean and Eurosiberian biogeographical regions. This mountain area is characterized as a hotspot of biodiversity, with numerous tree species and types of vegetation within a relatively small area (Longares Aladrén 2004). The main forests of the park are formed by European beech (*Fagus sylvatica* L.), Scots pine (*Pinus sylvestris* L.) and Pyrenean oak (*Quercus pyrenaica* Willd.), but there are other tree species present, such as Holm oak (*Quercus ilex* L.), Sessile oak (*Quercus petraea* (Matt.) Liebl.), Mountain pine (*Pinus uncinata* Ram.), European black pine (*Pinus nigra* Arnold.) and birch (*Betula pendula* Roth.). The closed-canopy forests are located in the northern half of the park; the southern area is dominated by shrubs, open-canopy *Quercus ilex* forests and small stands of other species.

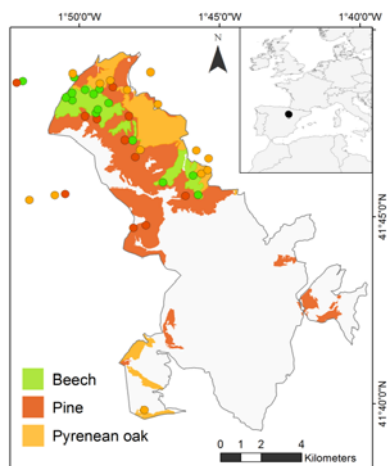


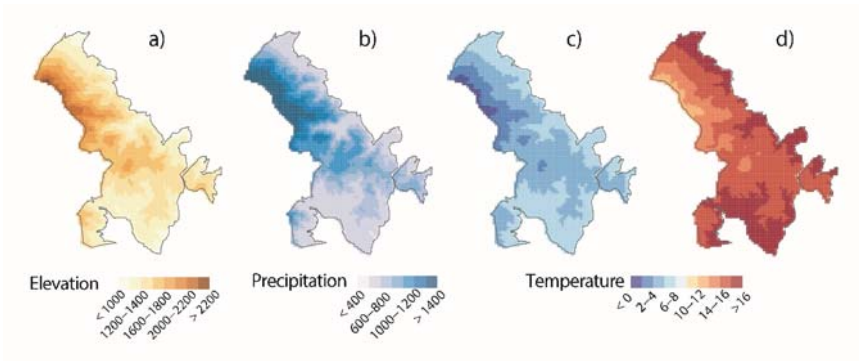
Fig. 1. Study area with current distributions (polygons) of study species and sample sites (circles) from which tree-ring chronologies were sampled inside and outside of Moncayo Natural Park.

2.2. Climate

The Natural Park is located in the eastern side of the Moncayo massif. Its geographical position represents an orographic barrier where the western and northwestern fronts prevent moisture from reaching the mid Ebro Valley. This situation produces a disruption of rising moist air flow that contributes to relatively frequent convective activity and results in higher amounts of precipitation at high elevations with a progressive decrease toward the lower elevations, whereas temperature follows the inverse path, with lower values at high elevations and higher values at low elevations.

With the aim to analyze the climate of the Moncayo Natural Park, three high-resolution gridded datasets of daily precipitation and maximum and minimum temperature were created using the *reddPrec* R package (Serrano-Notivol et al. 2017b) based on the reconstructed stations of Serrano-Notivol et al (2017a). The grids were based on a set of 2801 points regularly distributed over the study area with a spatial resolution of 200 m. Precipitation and temperature were computed daily for the period from 1950 to 2012. The annual averages for both precipitation (Fig. 2.b) and temperature (Fig. 2.c,d) show a clear altitudinal spatial distribution. The annual amounts of precipitation reach 1,500 mm at high elevations and decrease until values drop below 400 mm at lower elevations in the eastern and southern margins (Fig. 2b). The temperatures widely vary from the mean annual minimum below 0 °C at the summit to the mean maximum over 16 °C at the lower areas (Fig. 3). Annual temperature values show a similar pattern along the altitudinal bands with a change of -0.517 and -0.523 °C / 100 m increase in altitude for

158 minimum and maximum temperatures, respectively, whereas annual precipitation
159 increased by 119.89 mm / 100 m increase in altitude.



160
161 Fig. 2. Geographic and climatic gradients across study area for a) elevation, b) mean
162 annual precipitation, c) mean annual minimum and d) mean annual maximum
163 temperature.

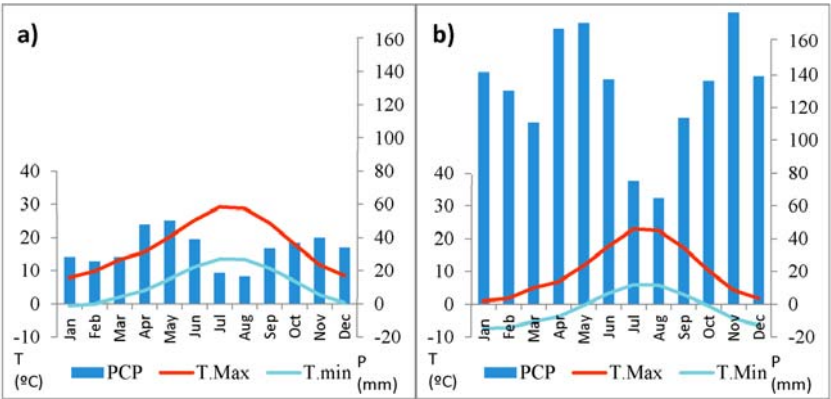


Fig. 3. Climatic diagrams at Moncayo Natural Park showing monthly precipitation (PCP), maximum temperatures (T.Max) and minimum temperatures (T.min) for a) lowest elevation 850 m.a.s.l. and b) highest elevation 2314 m.a.s.l.

2.3. Tree-ring chronologies

A total of 40 sites were selected and sampled between 2010 and 2014 along an altitudinal gradient from 950 to 1900 m.a.s.l. (Table 1, Table S1 in Supplementary Materials). Between five and 25 healthy and dominant trees were sampled at each site with an increment borer at breast height, extracting two cores per tree. Tree cores were mounted on wooden supports, dried, sanded and scanned in the laboratory. The tree rings were measured with the TSAP-Win program and LINTABTM 5 measuring device (Rinntech, Heidelberg, Germany) with 0.01 mm precision. Crossdating was done using Coorecorder v8.0 software (Cybis Elektronik & Data AB, Saltsjöbaden, Sweden).

Table 1. Summary of study sites for each species.

	No. sites	Altitude range	No. trees	Chronology range
<i>Fagus sylvatica</i>	15	1150-1600	150	1799-2014
<i>Pinus sylvestris</i> and <i>Pinus uncinata</i>	11	1020-1900	147	1919-2014
<i>Quercus pyrenaica</i>	14	950-1500	130	1875-2013

Ring-width measurements were detrended to remove biological trends using a negative exponential curve. A second detrending was then applied using a 50-year cubic smoothing spline with a 50% frequency response to filter out low frequency variation that reflects potential non-climatic disturbances. To obtain the detrended residual chronology to perform the analysis, two processes were done: an autoregressive modeling of the residuals and a bi-weight robust estimation of the mean. The detrending

process and chronology computation was carried out using the dplR package (Bunn 2008).

2.4. Statistical analysis

Correlation function coefficients (CFCs) were calculated between all 40 residual chronologies through seasonal 3-month means for respective maximum and minimum temperatures and precipitation sums from the previous summer (June, July and August) and the current summer. Correlation analysis was performed separately for each chronology with the associated climate calculated for each location.

As a second step, the variability and patterns of distribution of obtained seasonal CFCs for temperature and precipitation were explored. Generalized Additive Modelling (GAM) was used to detect relationships between the distribution patterns of obtained CFCs (dependent variable) and the mean annual values for precipitation and maximum and minimum temperature calculated for each study site (independent variables). This study is species-specific, but *P. sylvestris* and *P. uncinata* were grouped together due to similarities in climate-growth relationships found in this area between this species when compared with the other analyzed species. However, this similar behavior should not be extrapolated to other locations where both species are present.

The GAMs were constructed using the ‘gam’ package in R environment (Hastie and Tibshirani 1986). The quasi-binomial family was used in order to describe the error distribution (Eq 1). CFC values can theoretically vary from -1 to 1, therefore, observed values were re-scaled (CFC_{sb}) for model construction (Eq 2).

$$\text{gam}(CFC_{sb} \sim PCP + TMAX + TMIN + c(PCP:TMEAN), \text{family} = "quasibinomial")$$

(Eq 1)

$$CFC_{sb} = (CFC_s + 1) / 2 \quad (\text{Eq 2})$$

The accuracy of the models was evaluated using a likelihood ratio test by comparing the obtained models (full models) with restricted models where the explanatory variables of interest were omitted and only the intercept term was included (null models). The p-values for the likelihood ratio tests that compared the full and reduced models were calculated using the Chi-square distribution. In addition, explained variance (r^2) for each model was computed.

Finally, the obtained models were applied to the current species distribution in the MNP and also for the area of the Natural Park designated by a specific altitudinal range for each species. The boundaries of this altitudinal range are limited by the highest and lowest presence of a given species inside the Natural Park (from 1100 to 1750 m for beech; from 950 to 2100 m for pines and from 900 to 1500 m for Pyrenean oak). These areas are considered as potential distribution areas for the study species as they contain the specific environmental conditions required for their survival. The climatic conditions inside the different altitudinal gradients defined for each species are gathered in the sampled zones, therefore, the application of the models to these specific climatic conditions are not an extrapolation outside the values domain on which the models are constructed.

This approach is similar to the fundamental niche concept, where a combination of environmental variables defines the limits for survival without taking into account biotic competition (Reed and Clark 1978).

3. RESULTS

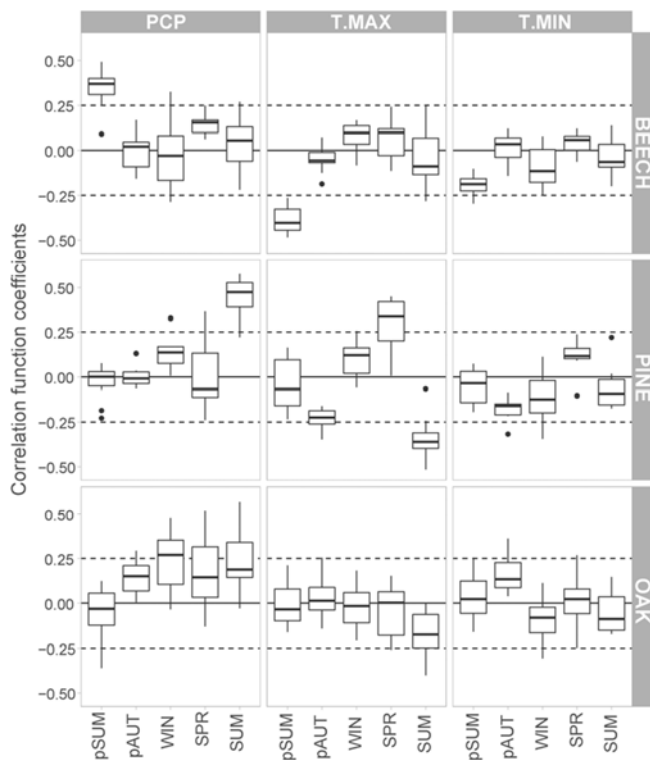
3.1. Chronologies

In total 427 trees were sampled in the area, covering an altitudinal range from 950 m.a.s.l. to 1900 m.a.s.l. (Table S1 in supplementary material). The youngest and oldest trees sampled for each species were 71 and 215 years for beech, 41 and 95 years for pine, and 37 and 271 years for Pyrenean oak. In total 40 chronologies were constructed, 15 for beech, 11 for pines and 14 for Pyrenean oak. The statistical analysis indicated a high quality of all chronologies based on the four commonly used statistical indicators in dendrochronology: the mean correlation between ring-width series (\bar{R}), the expressed population signal (EPS), the signal to noise ratio (SNR) and the mean sensitivity (SENS) (Table S1). \bar{R} values ranged from 0.21 to 0.55, with a mean of 0.42. SNR ranged from 1.33 to 28.20, with a mean of 12.41. SENS ranged from 0.2 to 0.46, with a mean value of 0.3. All EPS values reached the threshold of 0.85, indicating that all chronologies reflect a high common environmental signal of the sampled tree populations.

3.2. Climate-growth analysis

The results of the correlation analyses between the 40 chronologies and the climate variables (precipitation, maximum and minimum temperature grouped by seasons) by species are summarized in Fig. 4. The boxplots represent the variability of the correlation coefficients of all sampled sites of each species, therefore, across the studied populations (Fig. 4). If the box is completely beyond the dashed line (e.g. CFC for beech regarding T.max in previous summer), it implies that in all chronologies existed a significant relationship with the climate variable (in this case negative). If the box is completely inside the range described by the dashed line (e.g. CFC for beech regarding T.max in previous autumn), it implies that all chronologies do not showed a significant relationship with that climatic variable. If the box cross the dashed line (e.g. CFC for beech regarding T.max in winter), it implies that at least one chronology presented a

257 significant relationship (in this case, one positive and another negative, but in most of
 258 the cases the significance threshold was not reached). Lastly, the dashed line indicates
 259 statistical significance ($P < 0.05$) beyond 0.25 or -0.25, threshold given by the number
 260 of years included in the analysis (i.e. 62).



261
 262 Fig. 4. Climate-growth relationships for chronologies of beech (n=15), pines (n=11) and
 263 Pyrenean oak (n=14); dashed horizontal lines indicate statistical significance ($P < 0.05$)
 264 for 62 years of record; PCP = seasonal precipitation, TMAX, TMIN= seasonal
 265 maximum and minimum temperature, respectively. The central lines of boxplots
 266 indicate the median value, vertical hinges indicate first and third quartiles, error bars
 267 indicate the 95% confidence interval of the median and dots indicate outliers, values
 268 beyond the 95% confidence interval threshold.

269 Beech growth was significantly related to wet and cold conditions during the summer
270 preceding the current growing season. Precipitation had a positive and significant effect
271 in all but one case, meanwhile the maximum temperature affected growth negatively
272 (CFCs from -0.26 to -0.47). Large variation in correlation coefficients for current
273 summer (precipitation and maximum temperature) and especially winter conditions
274 (precipitation and minimum temperature) determine different responses among the 15
275 beech populations.

276 The growth of the pine populations was mainly driven by precipitation and maximum
277 temperature during summer in the current growing season with above-average rainfall
278 and below-average maximum temperature significantly favoring pine growth, whereas
279 during spring, high temperatures had a significant positive influence at the majority of
280 the pine sites. Large variation between pine populations occurred in correlations with
281 spring precipitation and maximum temperature during the previous summer.

282 Pyrenean oaks were especially sensitive to precipitation. There was a positive influence
283 of precipitation during all growing periods, even from the previous autumn. There was a
284 general negative influence of summer maximum temperatures, significant at some sites,
285 and a positive relationship between previous autumn minimum temperatures and
286 Pyrenean oak growth. In all cases, minimum temperatures had less influence on growth
287 than precipitation or maximum temperatures. Generally, the Pyrenean oak populations
288 vary more in their climate response than those of beech and pine as indicated by the
289 large size of the boxplots, although only the median correlation with winter
290 precipitation was significant.

291 The correlation analyses indicate large differences between the average response of
292 beech, pines and Pyrenean oak to prevailing climatic factors and, moreover, contrasting

293 responses within species to climate conditions across the environmental gradient of the
 294 Moncayo Natural Park.

295 3.3. Model application

296 To explore the variability of the sets of correlation coefficients yielded for the study
 297 species along the climate gradients for mean annual precipitation and the mean of the
 298 annual maximum and minimum temperatures within the study area, a total of 45 GAMs
 299 were constructed (5 seasons * 3 climate variables * 3 genera) and applied to the current
 300 species distribution in the MNP and also for the area of the Natural park designated by a
 301 specific altitudinal range for each species (considered as potential distribution areas).
 302 Predicted correlation coefficients for all analyzed seasonal climatic factors are shown in
 303 Fig. 5 (beech), 6 (pines) and 7 (oak). The reliability and validity of the models were
 304 evaluated based on the deviance and the significance level (Table 2).. Of the 45 models,
 305 22 were statistically significant.

306 Table 2. Deviance and significance of GAM relating CFCs variability and mean climate
 307 conditions in the study sites. ns – not significant ($p \geq 0.05$); * ($p < 0.05$); ** ($p < 0.01$);
 308 *** ($p < 0.001$).

309
 310

		BEECH		PINE		OAK	
		Deviance	Significance	Deviance	Significance	Deviance	Significance
PCP	pSUM	-0.1163	**	-0.0546	ns	-0.1291	ns
	PAUT	-0.1147	***	-0.0171	ns	-0.0730	ns
	WIN	-0.4489	***	-0.1018	***	-0.2105	ns
	SPR	-0.0275	ns	-0.3091	*	-0.3458	ns
	SUM	-0.2713	***	-0.1349	**	-0.1753	ns
	pSUM	-0.0724	***	-0.1343	ns	-0.1519	*
	PAUT	-0.0377	ns	-0.0231	ns	-0.1062	*

Con formato: Inglés (Reino Unido)

T.MAX	WIN	-0.0545	***	-0.0723	*	-0.1088	ns
	SPR	-0.1000	**	-0.2223	**	-0.1614	ns
	SUM	-0.3159	***	-0.1251	ns	-0.1115	ns
T.MIN	pSUM	-0.0281	ns	-0.0813	*	-0.1271	ns
	PAUT	-0.0645	***	-0.0251	ns	-0.0693	ns
	WIN	-0.1347	***	-0.1815	ns	-0.1386	*
	SPR	-0.0221	ns	-0.0887	ns	-0.0381	ns
	SUM	-0.1116	***	-0.1355	***	-0.1265	***

Fig.5. Beech predicted correlation coefficients for the previous summer (pSUM) and current summer (SUM) precipitation (PCP) and maximum temperature (T.MAX) across the MNP potential distribution area.

Fig. 6. Pine (*P. sylvestris* and *P. uncinata*): predicted correlation coefficients for spring (SPR) and summer (SUM) precipitation (PCP) and maximum temperature (T.MAX) across the MNP potential distribution area.

Fig. 7. Oak: predicted correlation coefficients for winter (WIN), spring (SPR) and summer (SUM) precipitation (PCP) and summer maximum temperature (T.MAX) across the MNP potential distribution area.

In beech, previous summer precipitation influence was positive and the maximum temperatures played a negative role across the potential distribution area (Fig. 5). The variations of CFCs associated with these climatic variables are not explained by the existing climatic gradients in the territory; therefore, the models are not significant (Fig. 2). Current summer conditions had a distinct influence across the territory, shifting from

positive to negative depending on altitude. The effect of precipitation for the current summer was positive in warm, dry, low-altitude areas, whereas it was negative at high altitudes where conditions were colder and wetter. High temperatures during summer affected growth positively at high altitudes and negatively at lower altitudes. Variations in CFCs during summer can be explained by climatic variability across the study areas as shown in Fig. 2.

Pine growth was influenced mostly by spring and summer conditions during the current growing season (Fig. 6). Precipitation in spring increased growth in low elevation zones, whereas in summer it was important all over the territory, especially in low elevation zones which are dryer. Maximum temperatures in spring increased growth in high elevation zones but in summer it became a growth-limiting factor, particularly in lower zones. In this case, CFCs significantly varied across climate gradients (Table 2). The amount of precipitation boosted oak growth during the entire year of the growing season, although with different intensity (Fig. 7). In general, the relationship between precipitation and growth was stronger in low elevation zones during spring and summer (Table 2). The occurrence of high temperatures during summer limited tree growth generally, with small variations across the study area.

4. DISCUSSION

4.1. Climatic influence on tree growth

In Moncayo Natural Park, the studied trees (beech, pines and oaks) showed different patterns of climate sensitivity. Similar observations in climate-growth relationships at different elevations in mountain areas have also been detected in other studies (i.e., Hartl-Meier et al., 2014; Kharal et al., 2017; Ponocná et al., 2016; Wang et al., 2017). Moreover, there are some common elements for all the species, such as the general importance of summer conditions for tree growth and the variation in correlation with specific climate factors along spatial gradients.

Beech growth has proven to be sensitive to drought at numerous sites (Weemstra et al. 2013; Tegel et al. 2014; Cavin and Jump 2016; Farahat and Linderholm 2018) and also in southern marginal stands in Spain (Rozas et al. 2015; Dorado-Liñán et al. 2017). In this study we found that wet and cold conditions during the previous summer are favorable for beech growth whereas the influence of current summer conditions varied significantly along the spatial climatic gradient. At lower elevation sites, beech growth is negatively influenced by high temperatures and low precipitation during summer, which supports results of other studies (Chen et al., 2015; Zimmermann et al., 2015) and can be explained by the species' strategy of rapid stomata closure to prevent the risk of embolism and inactivation of the water transporting system (Aranda et al. 2000; Granier et al. 2007). Conversely, at higher elevations, growth is enhanced when summer conditions are warmer. Trees may benefit from higher radiation on summer, that promotes photosynthesis in higher and colder conditions, that compensates water stress and embolism on beech trees. Čufar et al., (2008) described a negative influence of August maximum temperatures on beech growth at higher elevation sites in Slovenia.

374 This might be due to a prolonged growing season for the Slovenian beeches (until
375 August) as reported by Prislan et al., (2013) whereas at our study site the growing
376 period is rather short, lasting for 67 days at lower elevations and 80 days at higher
377 elevations, starting in May or early June and ending by late-July or early-August
378 (Martínez del Castillo et al. 2016). Therefore, beech may already be close to dormancy
379 during the late summer. The influence of winter precipitation on beech growth differs
380 along the altitudinal gradient; it is negative in high elevations and positive at low-
381 elevation sites.

382 The growth of pines was mainly favored by wet and cold conditions during the current
383 summer. In cold environments, high maximum temperatures during spring promoted an
384 early start to the growing season which leads to an extended growing season that
385 favored tree growth. Earlier cambium re-activation in pines is shown by heating
386 experiments (Gričar et al. 2007). In summer, this relationship changed and pine growth
387 was limited by high maximum temperature, which supports the findings of other studies
388 (e.g., Martínez del Castillo et al., 2018; Cailleret and Davi, 2011; Di Filippo et al.,
389 2007) and likely related to increased drought stress caused by high evaporation rates.
390 Pines are known to exhibit a plastic growth behavior which is indicated by their ability
391 to anticipate favorable spring conditions and start growing early and also by re-
392 activation of their growth after cessation from dry conditions during summer (de Luis et
393 al. 2007; Camarero et al. 2010; De Luis et al. 2011; Novak et al. 2016). Furthermore, at
394 southern and/or low elevation sites with mild temperatures during winter, pines may not
395 enter in a real dormancy period (Gričar et al. 2016; Prislan et al. 2016).

396 In the case of oak, water availability is largely determined by the amount of
397 precipitation, which triggers tree growth in lower and dryer areas. Although oaks are
398 known to be able to recover water potential due to a deeper and extensive root system

when compared with beech (Aranda et al. 2000), a severe and prolonged water deficit leads to an irreversible cavitation of earlywood vessels with tylose formation (Pérez-de-Lis et al. 2018). Several studies identified the positive effect of spring/summer precipitation on the growth of ring-porous species (García-Suárez et al. 2009; García-González and Souto-Herrero 2017). Similarly, high maximum temperatures during summer limit oak growth across sites in Central Europe (Čufar et al., 2014; Mérian et al., 2011).

4.2. Spatial distribution of correlations between climate and growth

As hypothesized, distribution patterns for correlations between growth of tree species and prevailing climate factors are at least partially, related to the existing climate gradients. The significance of the models proved that these influences of the seasonal climatic variables are explicable by the climate variability over the territory. This fact is most noticeable in beech, with 11 significant models out of 15, somewhat in pines, with 7 significant models out of 15 and less evident in oak, with only 4 significant models out of 15. Climate-growth relationships for the three species were tested with GAMs to create models which allow projection of the influence of climate on the growth of selected species along climate gradients within the study area.

Although the variance of some variables was largely unexplained and statistical significance was lacking (i.e., previous summer precipitation in beech), information obtained from the GAMs is still valuable. On one hand, significance models suggest the existence of a plastic climate-growth relationship for the species across the studied distribution area. Such plasticity indicates that the influence of a given climate factor is climatically determined and would probably be modified if changes in climate occur. This is key for the development of accurate predictions of species behavior across a

study area under different climate change scenarios. Another fact is that the lack of significant models suggests that the influence of a given seasonal climate factor is constant across the study area (or at least not affected by prevailing climate conditions). The influence of such a seasonal climate factor is then expected to be more stable if changes in climate occur (at least if such changes do not exceed the actual climate range within the study site). *4.3. Applicability of the model approach in Moncayo Natural Park*

Over the last several decades, forest-management actions performed in MNP aimed at conserving and improving forested habitats (Martínez del Castillo et al. 2015). Specifically, the objectives for beech and Pyrenean oak forests are to improve their natural regeneration and to improve the quality and status of conservation, since these species are protected by the Habitats Directive. Conversely, pines were introduced in the park as reforestations and they are not protected by the Directive.

Our results showed that beech forest may be favored in the middle of the actual altitudinal range (1400-1500 m.a.s.l), and progressively extend to southern locations in the center of the Park where the species is currently not present. The climatic conditions of this new area are similar to the climatic conditions where beech forest is currently growing, and our models forecasted similar climate-growth relationships for the most important climatic variables influencing beech growth. However, it could not spread in the lowlands of this mountain due to higher temperatures and water scarcity.

The specific management actions in MNP for pines are aimed to naturalize the structure of the forest and reduce density by low-impact logging. The pine forest models showed that forest could be extended to high zones in the center part of the Park, which is

partially unforested, given the similar climatic conditions of that area and the models results. By spreading pine forest cover to other areas, with logging actions the current pine forest at medium altitudes could be progressively replaced by protected species, in this case beech or Pyrenean oak, avoiding at the same time pine loss at MNP.

Pyrenean oak forests are affected by drought especially in low zones; therefore, logging actions to reduce tree density during dry years could compensate for growth reduction by decreasing competition. In addition, Pyrenean oak forests in Mediterranean areas largely have over-aged coppice stands where growth is reduced and trees are more sensitive to summer drought leading to oak-decline (Corcuera et al. 2006). Therefore, management actions to create uneven aged stands, as the proposed logging, may become necessary. Our results showed a positive reaction to water availability, consequently, these actions may not be necessary in rainy years, but could be very helpful during dry years.

Finally, the models suggest that the lower elevations in the southern part of the Park, in general, are not suitable for these tree species due to slightly different climate conditions.

4.4. Potential use of the model approach as a forest-management tool

The climate-growth models, as presented in this study, provide a novel and complementary approach to forest-growth models or species-distribution models, which are important tools for forest management (Porté and Bartelink 2002; Wullschleger et al. 2014). However, those models are based on data mining or regression techniques and hence largely lack ecophysiological explanations (Walentowski et al. 2017). Although the present and future distribution of tree species is a complex combination of multiple biotic and abiotic factors, to date, models based on climate provide the best available

guide for policy making (Pearson et al. 2004; Booth 2017). Moreover, our models are applied to a simple version of the fundamental niche for each species, unlike most of the species distribution models, that are based on current distribution without taking into account information on species climatic requirements beyond their potential distributions (Booth 2017).

The models are empirical and therefore the forecasted CFCs are applicable to the environmental domain from where the samples were taken. In this sense, the model interpolates the results within the environmental gradient (climatic-altitudinal gradient in this case) of the sampled area. In addition, both CFCs analysis and the modelling were carried out in a specific range of climatic variability; therefore the results of our models are not able to be extrapolated outside this range. The extrapolation of the results outside the range (i.e., other locations of this forests species over Europe) would include unsolvable uncertainties that would produce unreliable results (National Research Council 2012). Nevertheless, the applicability of our new methodological approach is universal and may be applicable to any other environment. However, application of the approach requires a systematic sampling strategy to ensure that the application of the models do not exceed the range of values used to construct the models.

The same methodological approach can be applied to predict changes in climate-growth relationships under future climate-change scenarios. This requires input of a tree-ring network and high-resolution climate data, in addition to high-resolution future climate projections. Future climate conditions can exceed the range of the current climate variability; therefore, the models would have to extrapolate the results which might not be realistic. A methodological effort should be done in the future to address these topics, due to the importance of forest performance in future climates.

Forecasting future distribution of species is a complex task. The plastic response of species to climate variability is a key factor but it is not sufficiently incorporated into current modelling approaches. Complementary approaches, as presented here, can improve these models, but dense tree-ring networks are required.

5. CONCLUSIONS

The study on the growth response of beech, Scots pine, Mountain pine and Pyrenean oak in Moncayo Natural Park yielded information on spatial changes in climate-growth relationships across the altitudinal gradient. This was mainly related to different climatic conditions; therefore, the altitude of the sampling needs to be considered in dendroclimatological studies performed in the edge distribution limit of the species or in areas with large elevation gradient.

The study presents the current relationship between tree species and climate within their realized elevation range as well as the design and application of tree-ring based models to predict performance of the species across the potential distribution areas in the Park.

The model results indicate that the total forested area could be expanded with additional management actions, mainly to higher altitudes and in a central zone of the Park which is currently unforested.

Forest management policy needs better tools to manage the distribution of tree species on afforested areas according to current and expected changing climatic conditions. The newly developed modeling approach can form a key tool to support concrete forestry decisions about species selection and estimation of species performance in specific areas for which it is calibrated. Apart from their direct application, the models showed a potential for direct integration into decision-making systems in a sustainable multifunctional forest management environment. The application of the actions derived

from the models would enhance sustainable forest management and could help to mitigate climate change effects.

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718

719 Table S1

Site	Species	Altitude	Chrono	No.	RBar	EPS	SNR	SENS
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Code		length	Trees					
M01	<i>F. sylvatica</i>	1200	1902-2014	21	0.41	0.95	19.96	0.33
M02	<i>F. sylvatica</i>	1400	1829-2010	10	0.51	0.95	17.66	0.41
M03	<i>F. sylvatica</i>	1600	1832-2011	10	0.51	0.92	11.30	0.46
M04	<i>F. sylvatica</i>	1420	1835-2010	10	0.50	0.92	11.34	0.42
M05	<i>F. sylvatica</i>	1500	1912-2010	15	0.41	0.94	15.97	0.38
M06	<i>F. sylvatica</i>	1170	1914-2010	5	0.53	0.90	8.70	0.35
M10	<i>F. sylvatica</i>	1600	1799-2014	20	0.48	0.97	27.27	0.39
M11	<i>F. sylvatica</i>	1200	1904-2010	10	0.39	0.91	10.57	0.37
M14	<i>F. sylvatica</i>	1440	1949-2010	5	0.42	0.85	5.28	0.34
M16	<i>F. sylvatica</i>	1150	1902-2010	10	0.40	0.92	10.86	0.37
M17	<i>F. sylvatica</i>	1520	1887-2010	10	0.44	0.89	7.89	0.33
M26	<i>F. sylvatica</i>	1380	1825-2011	6	0.46	0.90	8.98	0.38
M27	<i>F. sylvatica</i>	1320	1830-2011	6	0.44	0.88	7.60	0.32
M28	<i>F. sylvatica</i>	1255	1836-2011	6	0.49	0.90	9.15	0.35
M29	<i>F. sylvatica</i>	1177	1904-2011	6	0.47	0.91	9.48	0.33
M01	<i>P. sylvestris</i>	1020	1919-2014	25	0.37	0.96	24.45	0.30
M02	<i>P. sylvestris</i>	1510	1938-2014	25	0.44	0.96	23.96	0.23
M03	<i>P. sylvestris</i>	1635	1952-2009	13	0.40	0.93	12.99	0.25
M04	<i>P. sylvestris</i>	1420	1923-2010	5	0.39	0.84	5.25	0.25
M07	<i>P. sylvestris</i>	1190	1933-2010	14	0.53	0.96	26.29	0.30
M13	<i>P. sylvestris</i>	1400	1951-2010	15	0.51	0.97	28.20	0.28
M17	<i>P. sylvestris</i>	1520	1964-2010	10	0.37	0.91	10.31	0.25
M18	<i>P. sylvestris</i>	1480	1972-2010	10	0.53	0.95	17.89	0.20
M19	<i>P. uncinata</i>	1780	1962-2010	10	0.35	0.89	8.03	0.25
M20	<i>P. uncinata</i>	1900	1969-2010	10	0.26	0.85	5.52	0.25
M40	<i>P. uncinata</i>	1765	1935-2011	10	0.38	0.91	9.62	0.21
M06	<i>Q. pyrenaica</i>	1190	1968-2010	5	0.30	0.88	3.60	0.23
M12	<i>Q. pyrenaica</i>	988	1940-2010	12	0.50	0.94	16.52	0.24
M13	<i>Q. pyrenaica</i>	1400	1951-2010	14	0.43	0.95	18.77	0.31
M16	<i>Q. pyrenaica</i>	1150	1968-2010	10	0.34	0.90	8.58	0.24
M18	<i>Q. pyrenaica</i>	1550	1895-2010	10	0.29	0.86	2.58	0.25
M21	<i>Q. pyrenaica</i>	1320	1869-2010	5	0.27	0.85	1.33	0.27
M23	<i>Q. pyrenaica</i>	950	1973-2011	10	0.45	0.92	12.16	0.25
M24	<i>Q. pyrenaica</i>	1020	1974-2011	10	0.55	0.95	20.97	0.33
M25	<i>Q. pyrenaica</i>	1050	1968-2011	9	0.37	0.91	9.97	0.22
M30	<i>Q. pyrenaica</i>	1154	1950-2011	6	0.49	0.86	6.09	0.24
M31	<i>Q. pyrenaica</i>	1036	1966-2012	9	0.43	0.92	11.39	0.30
M32	<i>Q. pyrenaica</i>	1125	1963-2012	10	0.40	0.92	11.48	0.28
M33	<i>Q. pyrenaica</i>	1155	1950-2012	10	0.38	0.89	7.93	0.24
M34	<i>Q. pyrenaica</i>	1060	1967-2013	10	0.39	0.92	10.72	0.28

Figure S2. Beech predicted correlation coefficients.

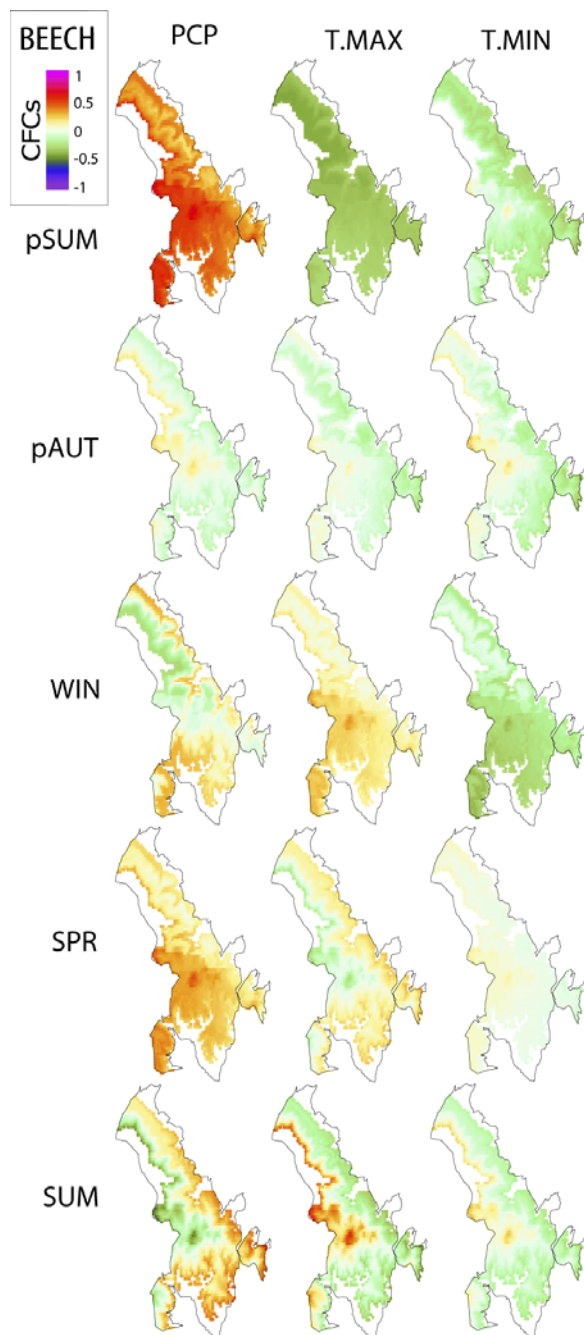


Figure S3. Pine predicted correlation coefficients.

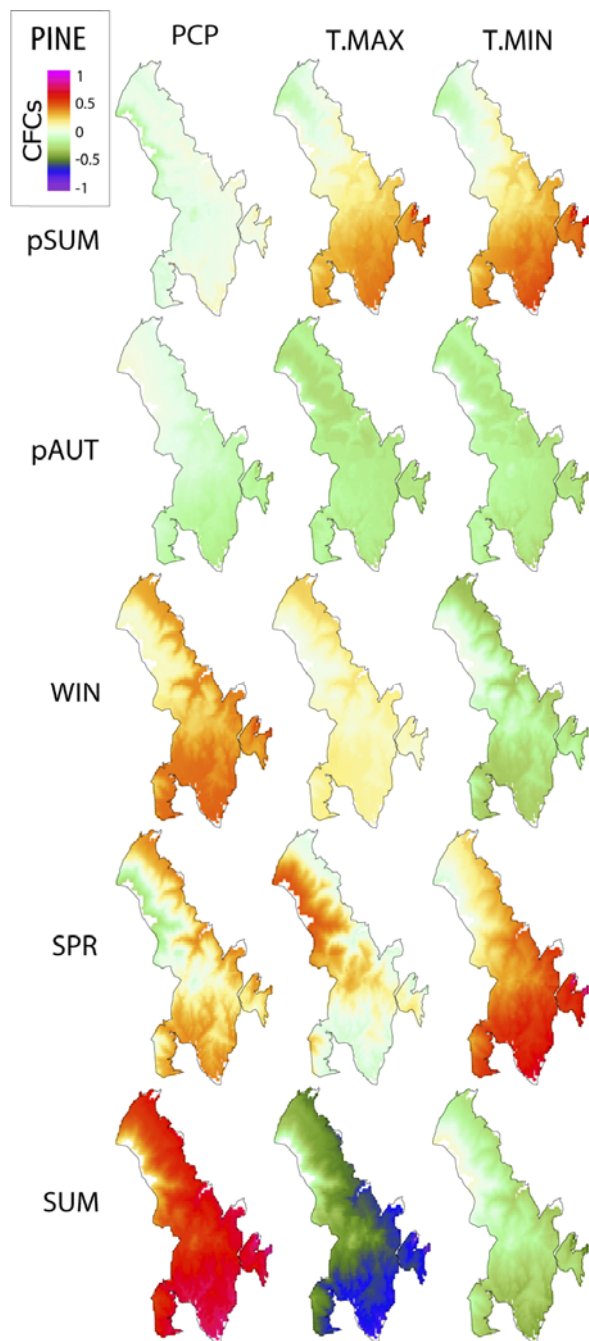


Figure S4. Pyrenean oak predicted correlation coefficients.

